

Simple and Fast Interval Assignment Using Nonlinear and Piecewise Linear Objectives

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International Meshing Roundtable October 2013

20 minute talk, 5 minute questions





Outline

- Problem definition
 - a few simple examples of why it is hard
 - · tradeoffs needed
 - · may be infeasible
 - an infeasible example
- Prior approaches
 - Linear constraints good
 - Linear objective not so much IMHO, depending on setting
 - Tam & Armstrong concentrates deviations
 - · Lex min max is slow series of problems
- New solution
 - Cubic Objective
 - Piecewise linear to get integer solution
 - Implemented

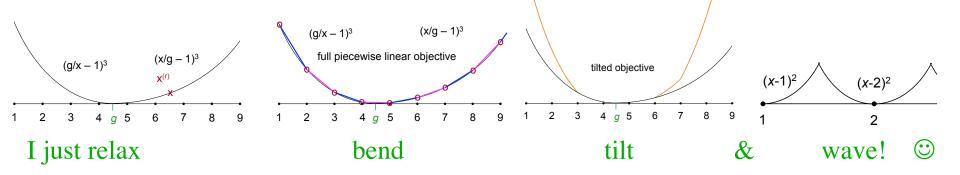


Interval Assignment Summary

Simple and Fast Interval Assignment Using Nonlinear and Piecewise Linear Objectives

Scott A. Mitchell

- New nonlinear objective function
- Finesse integer constraints using $\min f(x)$ L₁ minimization of convex piecewise linear approximation $\lim_{x_I \in \mathbb{N}} x_I = 0$
 - adapt piecewise, slope $l \le x \le u$.
 - · heuristics for constraint interactions

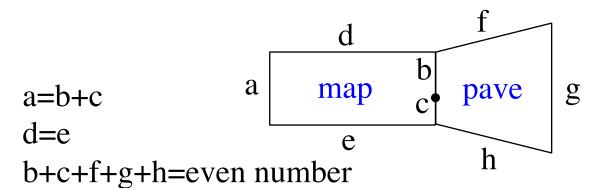


- First known improvement to FEM interval assignment structure since 1997 lexicographic min-max
 - Lots of people write new constraints for new schemes, but this is first change to objective (except Graphics community solving related-but-different problem using MIQP)
- Implemented in MeshKit, using IPOPT
 - scales to thousands of surfaces in nuclear reactor core models
 - 1997-2013, Cubit runs lex min-max for every quad and hex mesh
 - New solution may migrate to Cubit, as Cubit includes MeshKit library



Interval Assignment (IA) Problem Definition

- Set the number of mesh edges on each curve, such that the owning surfaces and volumes can be meshed according to their schemes.
 - compatible interface meshes, e.g. for parallel
 - quad and hex meshing
 - triangle and tet meshes have trivial constraints

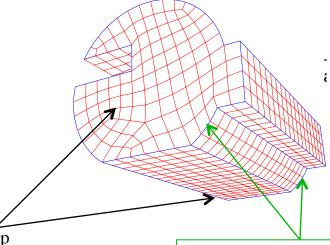


all variables integer (e.g. a=4, b=2.5, c=1.5 is not valid)



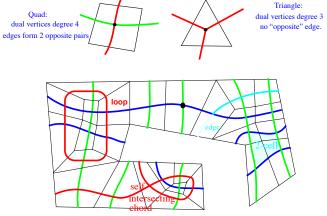
Why is Interval Assignment Hard?

- What's the problem?
 - hex meshing is a global problem
 - quad meshing is a global problem
 - for the same reasons,
 - but with much weaker constraints
- Integrality makes it much harder.
 - Discrete problem
 - Combinatorial algorithms needed



Sweep same quad mesh connectivity front and back

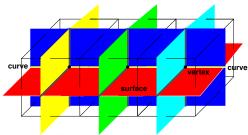
Map same curve mesh connectivity top and bottom. e.g. 3 edges



A quadrilateral mesh and the corresponding STC.

Unstructured quad meshing:

Each chord enters the quad mesh an even number of times (0 or 2) so we need an even number of mesh edges on the boundary. (necessary and topologically sufficient)



Unstructured hex meshing:

Each chord enters the hex mesh an even number of times (0 or 2) so we need an even number of quads on the boundary.

Plus contractible loops must be even. (necessary and topologically sufficient)



Interval Assignment by linear constraints is old

Interval assignment problem form

- · Linear constraints, linear objective
 - Tam & Armstrong 1993
- Linear constraints, lexicographic min-max objective
 - Mitchell 1997
- Fluid flow matching
 - Muller-Hanneman 1997
- Linear constraints, quadratic objective
 - Bommes et al. 2009+

Interval assignment constraints

- Volumes with holes
 - Shepherd et al.
- Midpoint subdivision
- Paver
- Submapping
- Skew control

Automatic Scheme selection

White & Tautges 2000

Popular to define constraints for new meshing schemes.

Little change to *objective function form* in FEM community 1997—this paper

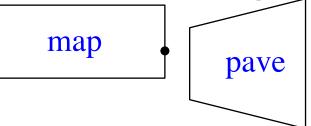
(But active area in Graphics 2009+)



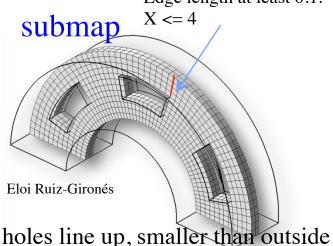
Local Constraints

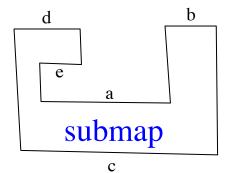
curve has at least one edge x >= 1

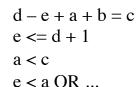
- Each surface has local constraints
 - depending on scheme (structure)
 - depending on user desires (sizing, skew)

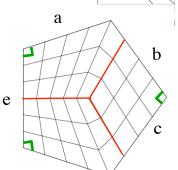


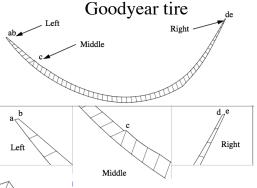
Edge length at least 0.1:











midpoint subdivision side1 = a+bside2 = c+d

side2 = c+dside3 = e

side1 <= side2 + side3

 $side2 \le side1 + side3$

 $side3 \le side2 + side3$



Mixed-Integer Linear Constraints

- Constraints are linear, no high powers such as $x^2 = y^3$
- Coefficients are typically 1.
- Sum-even modeled with artificial variable, 2-coefficient
 - $sum(x_i)=even \Leftrightarrow sum(x_i) = 2z : z is integer$
- Expressed in matrix form

$$\min f(x)$$
s.t. $Ax = b$

$$x_I \in \mathbb{N}$$

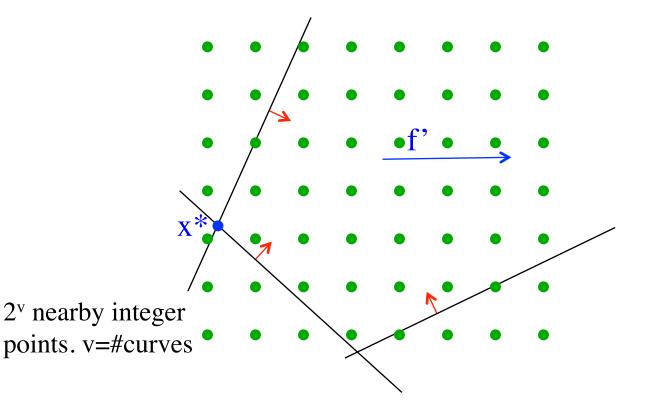
$$l < x < u.$$

- Inequality and equality are equivalent
 - $Ax ≥ b \Leftrightarrow Ax + y = b, y ≥ 0$
 - $Ax = b \Leftrightarrow Ax \leq b$ and $Ax \geq b$



Benefits of Linear Constraints

- Feasible region (space of solutions) is convex (possibly unbounded)
 we have efficient solution algorithms (especially for linear objective)
 - Not so easy for integer solutions



$$\min f(x)$$
s.t. $Ax = b$

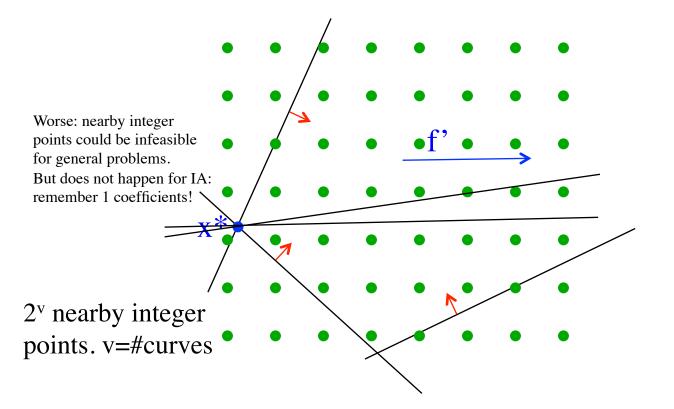
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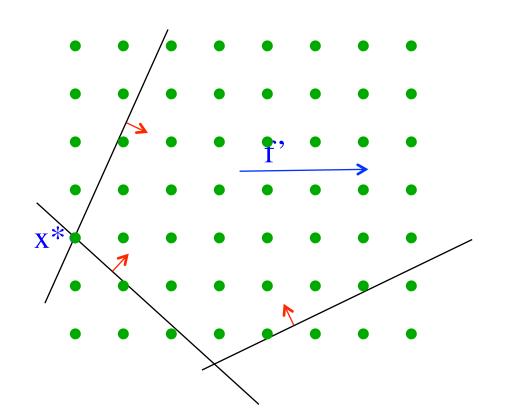
$$x_I \in \mathbb{N}$$

$$l \leq x \leq u$$
.



Benefits of Linear Constraints

- Feasible region (space of solutions) is convex (possibly unbounded)
 we have efficient solution algorithms (especially for linear objective)
 - Not so easy for integer solutions
 - Unless coefficients are cooperative!



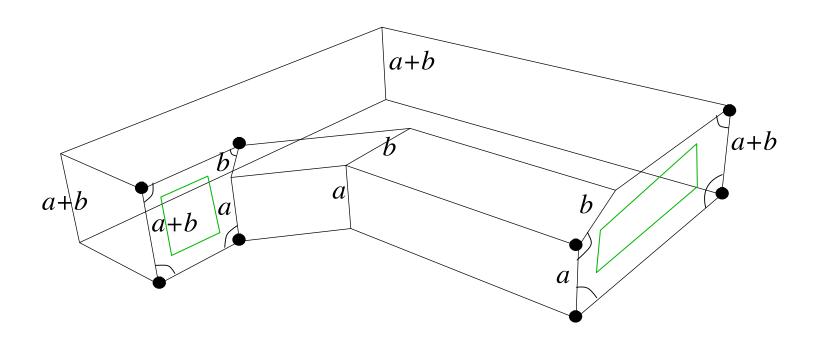
$$\min f(x)$$
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$$x_I \in \mathbb{N}$$

$$l \le x \le u.$$



Global Infeasibility



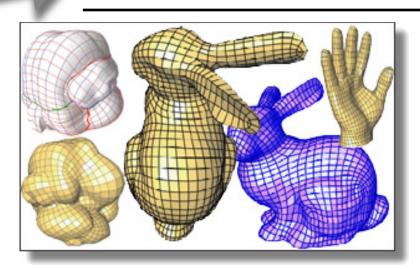
$$a = a + b, b = 0, x$$

The global IA problem may be infeasible, meaning no solution exists, even when each surface can be meshed in isolation.



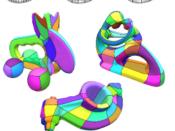
State of the Art in Quad Meshing **Eurographics STARS 2012**

Graphics has discovered quad meshes



Designing Quadrangulations with Discrete Harmonic Forms Yiving Tong, Pierre Alliez, David Cohen-Steiner, and Mathieu Desbrun ACM/EG Symposium on Geometry Processing 2006, pp. 201-210





Mostly structured quads on smooth curved surfaces Key issue is placement of irregular nodes

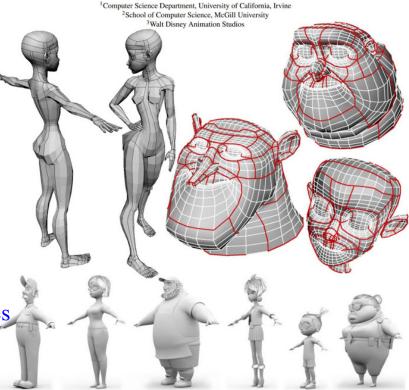
Integer-Grid Maps for Reliable Quad Meshing David Bommes, Marcel Campen, Hans-Christian Ebke, Pierre Alliez, Leif Kobbelt SIGGRAPH 2013

Motorcycle Graphs: Canonical Quad Mesh Partitioning

Eurographics Symposium on Geometry Processing 2008 Pierre Alliez and Szymon Rusinkiewicz

(Guest Editors) http://www.disneyanimation.com/library/motorcycle_sgp_2008.pdf

David Eppstein^{†1}, Michael T. Goodrich^{†1}, Ethan Kim², and Rasmus Tamstorf³



Disney Animation

(Pixar uses triangles)



Graphics has discovered quad meshes including the need for interval assignment

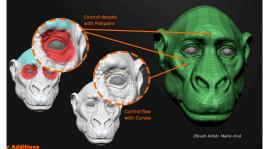
Solved with mixed-integer linear-constraint optimization, series of problems The placement of irregular nodes is included in the optimization problems



commercial version

Mixed-Integer Quadrangulation.
David Bommes, Henrik Zimmer, and Leif Kobbelt.
ACM Trans. Graph. (TOG), 28(3):77:1–77:10, July 2009.

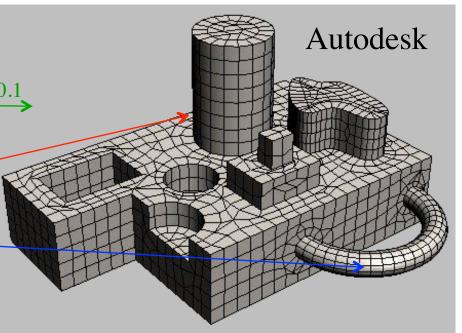






Objectives: user specified sizes

- Goal g for each interval
 - about g
 - each mesh edge should be about 0.1 inches long $\frac{g = length}{x \approx g}$
 - exactly g (not implemented currently)
 - I want exactly 8 intervals here-
 - at least g
 - I need at least 6 here $\frac{g=6}{x>a}$

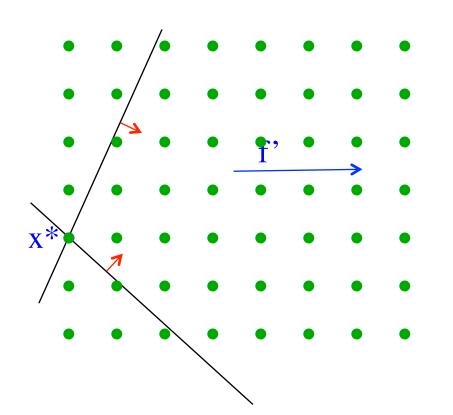


- Achieving all of these g, and constraints, may not be possible.
 - How do you measure deviations? Relative values?



What's good about linear objectives?

- Optimal solutions occur at corners, easy to find.
 - if constraint coefficients cooperate, corners are integer solutions. Integer solution for free!



$$\min f(x)$$
s.t. $Ax = b$

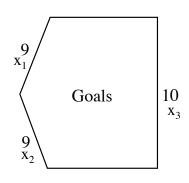
$$x_I \in \mathbb{N}$$

$$l \le x \le u.$$



What's wrong with linear objectives?

- All the deviations are concentrated
 - L₁ minimization leads to sparsity in the solution
 - not a big deal if deviations are small
 - could be a big deal if deviations are large



$$g_1 = 9$$

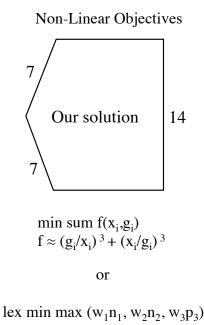
$$g_2 = 9$$

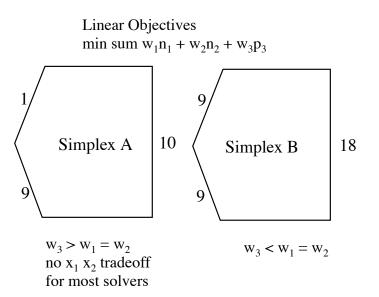
$$g_3 = 18$$

$$x_1 = g_1 - n_1$$

$$x_2 = g_2 - n_2$$

$$x_3 = g_3 + p_3$$





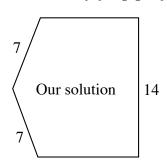


My Solutions

to spread out deviations

- 1997 Lexicographic min max solve series of linear programs, one per variable = slow.
 - Minimize maximum x_i - g_i . Fix that x_i at a nearby integer value, check feasibility, remove it from equations.
 - minimize maximum remaining x_i-g_i, repeat
 - Weighted deviations, relative size
 - min max $p_i(x_i-g_i>0) + n_i(g_i-x_i>0)$, where $p_i = 1/g_i$ $n_i = 1.3/g_i$
 - i.e. assigning x=10 for g=5, is as bad as x=20 for g=10
- Today, cubic objective (L₃)
 - A small series of nearby problems to get integer
 - About the same solution as lex min max, depending on problem size
 - L_∞vs. L₃ norm

lex min max (w_1n_1, w_2n_2, w_3p_3)



min sum
$$f(x_i,g_i)$$

 $f \approx (g_i/x_i)^3 + (x_i/g_i)^3$



Cubic Objective

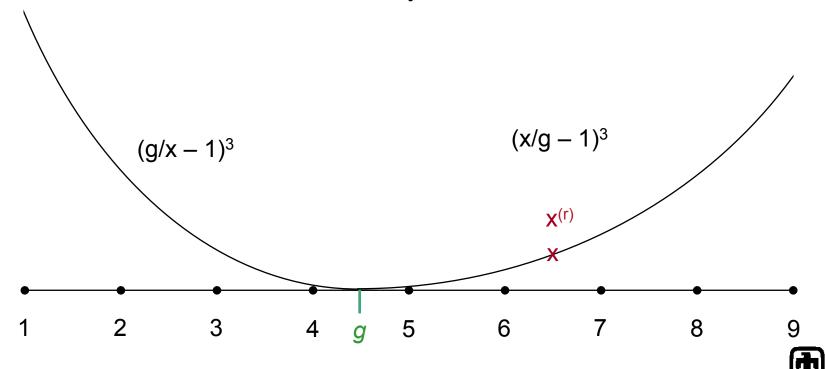
Good: directly measures relative change, (g->x: 4->8) is as bad as 4->2

Good: no lexicographic min max, single optimization solve spreads deviations

Good: convex objective (plus convex constraints)

local optimum is the global optimum

Bad: non-linear objective requires slower algorithms than a linear one, but the difference is less today than in 1997



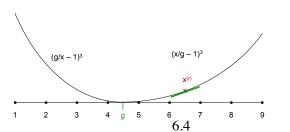
Cubic Objective Solution

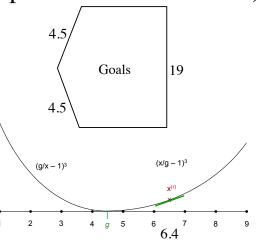
Non-integer solution usually

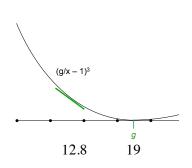
(later problem definitions fix)

min sum
$$f(x_i,g_i)$$

 $f = (g_i/x_i-1)^3$ or $(x_i/g_i-1)^3$

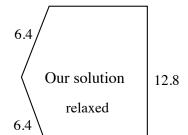






min F =
$$(x_1/4.5 - 1)^3 + (x_1/4.5 - 1)^3 + (19/x_3-1)^3$$

minimum when F' = 0 (or extremes $x_i=1$, etc.)
sum of slopes = 0

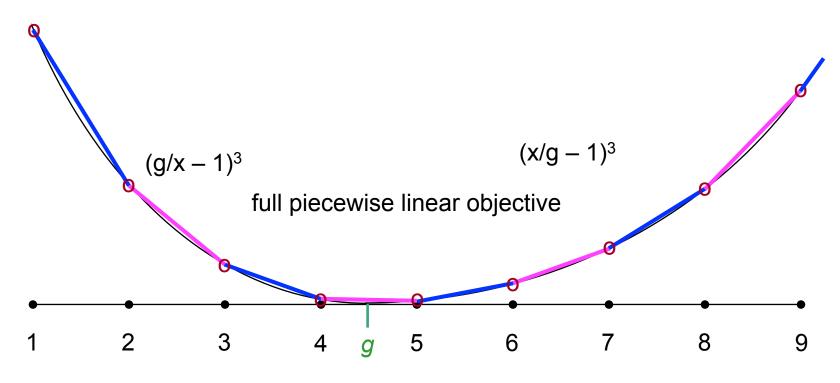


I have spread out the deviations, made the major compromises

Piecewise Linear Objective

Good: restores integer solutions for free (often)

Bad: explicit variable for every single unit interval = expensive time, memory

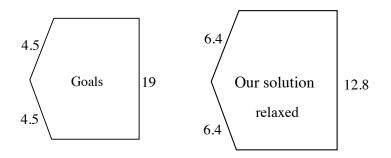


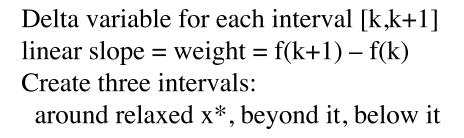
Bend at integer points <-> corner at every integer in the higher-dimensional linear problem where each unit interval is a dimension.



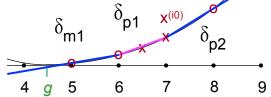
Adaptive Piecewise Linear Objective

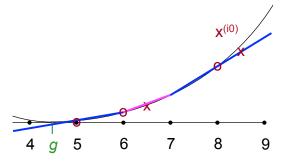
For efficiency, just add the bends (pieces) as needed

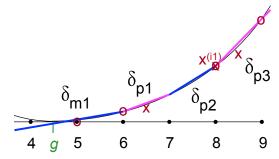


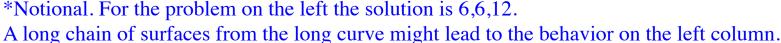


Solve min sum weighted deltas (linear) Add more deltas if solution beyond k+1, repeat







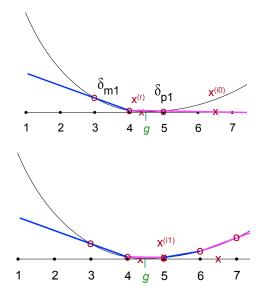




Adaptive Piecewise Linear Objective

Add the bends (pieces) as needed

Start with two bends, so relaxed solution is in a trough to avoid unbounded solutions right away





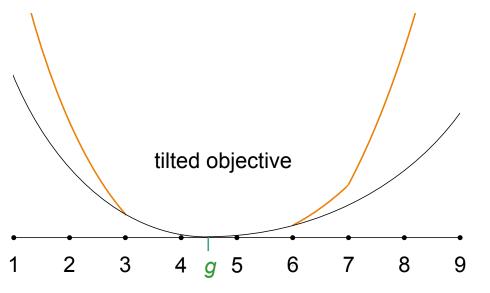
Tilt to Break Non-integer Ties

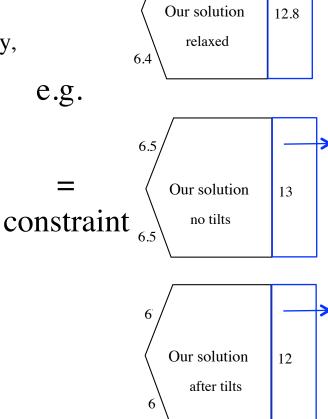
Break non-integer ties by tweaking the weights (of the deltas), an additive factor to the slope of the objective.

Randomization helps avoid cases where, say, many variables to the right gang up to motivate a non-integer solution on the left.

Scaling heuristics needed when goals or deviations vary widely, else small goal variables are left at non-integer values.

Numerical tolerance issue with solvers





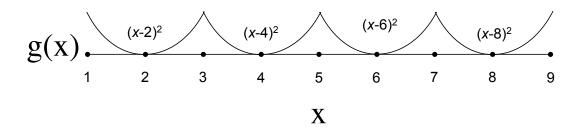


Last Resort, Waves to Force Integrality

unstuck ½ (fractional) integer values

e.g.

x=sum of edges, need an even number constrain g < 0.001



13 no tilts Our solution after waves

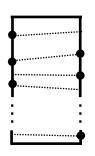
Our solution

Bad: breaks convexity, local minima are not global minima IPOPT has limited global optimization capabilities

Better, in progress: bends and tilts for sum-even variables





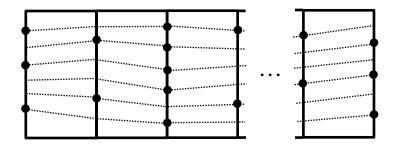


scaling by curves

curves	$x^{(r)}$ s	$x^{(r)}$ c/s	$x^{(i)}$ s	$x^{(i)}$ c/s	$x^{(i)}/x^{(r)}$ time	total s	total c/s
2,000	0.027	74k	0.084	24k	3	0.11	18k
6,324	0.051	120k	0.37	17k	7	0.43	15k
20,000	0.14	140k	1.9	11k	14	2.0	10k
$63,\!238$	0.48	130k	7.0	9k	15	7.5	8k
200,000	2.0	100k	50	4k	25	52	4k
632,378	11	57k	430	1.5k	39	444	1.4k



Runtime about $O(c^{1.5})$ for c variables.



scaling by faces

fa	aces	curves	$x^{(r)}$ s	$x^{(r)}$ f/s	$x^{(i)}$ s	$x^{(i)}$ f/s	$x^{(i)}/x^{(r)}$	$total\ s$	total f/s
1	160	1,061	0.028	5700	0.061	2600	2.2	0.091	1800
5	505	3,298	0.067	7500	0.21	2400	3.1	0.27	1900
1,	,600	10,614	0.23	7000	0.74	2200	3.2	0.98	1600
5,	,050	32,793	0.80	6300	2.3	2200	2.9	3.2	1600



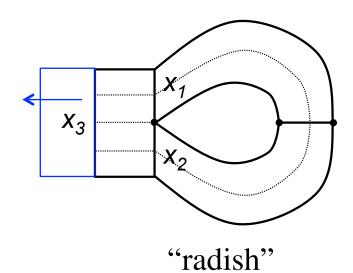
Time for an *integer* solution is only a small multiple of the time for the *relaxed* solution.

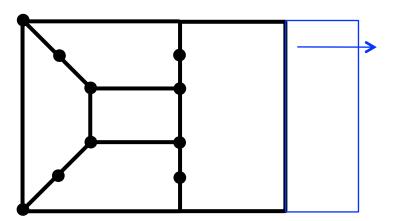
In contrast, Lex min max grows quadratic+ in problem size



Robustness stress tests

Problems needing tilts to get integer solutions





Split curves provide more degrees of freedom: more tilts are needed to remove fractional values.

Only a few tilts are necessary, tilt for each split curve in the chain is simultaneous.



Availability and Status

- MeshKit algorithm implementation
 - Open source
 - http://gnep.mcs.anl.gov:8010/meshkit/
 - Well defined API
 - IPOPT optimization library is free
 - MA27 linear algebra library is free
 - but you have to download it yourself
 - HSL (formerly the Harwell Subroutine Library). A collection of Fortran codes for large scale scientific computation. http://www.hsl.rl.ac.uk, 2011. Includes MA27. IPOPT special instructions at http://www.hsl.rl.ac.uk/ipopt/
- To do:
 - Robustness and quality testing on large models
 - Tilt for sum-even constraints
 - better than just waves

Main Page Related Pages Namespaces Classes Files Search for

%MeshKit: A Library for Mesh Generation

0 0

MeshKit is a library for geometry-based 2d and 3d mesh generation. This library has two main purposes:

- · to provide a useful collection of open-source mesh generation functions
- to provide infrastructure (geometry, lower-dimensional entity meshing, etc.) for implementing new meshing functionality

MeshKit uses a component-based approach, using external components for selected functionality where possible. Geometry and relations to mesh are accessed through the iGeom and iRel interfaces, resp. Mesh is accessed through both the MOAB and iMesh interfaces. MeshKit allows registration of external meshing algorithms, allowing those algorithms to operate in collaboration with the rest of MeshKit functionality.

This manual is divided into the following sections:

- · User's Guide
 - Using MeshKit: The 2-Minute Introduction
 - Data Model
 - A Graph-Based Model for Mesh Generation
 - A more detailed example of graph traversal
 - More Graph-Based Meshing Examples
 - Geometry, Mesh, Relations Interfaces
 - MeshKit API
 - MeshKit Algorithms
- Developer's Guide
 - · Coding Style Guide
 - · Geometry, Mesh, Relations Interfaces
 - Writing A New MeshOp Class
 - Frequently Asked Questions

Generated on Thu Mar 3 15:45:39 2011 for MeshKit by



Alternatives

- Why bother with the nonlinear solver?
 - it is fast, uses little memory
 - but maybe you don't have one
- Skip the relaxed cubic-objective phase
 - start with piecewise linear adaptive bends
 - efficiency would depend on numerics, how far the optimal solution is to the initial goals



Alternatives

- Why sum-of-cubes objective function?
 - Any power in [2, infinity] is possible.
 - Lex Min Max from 1997 is a form of L_infinity
 - The lower the power, the bigger the worst deviation, but the fewer number of curves need to change
 - Experimental. I tried powers of 2, 3, and 4.
 - 2 concentrated deviations too much
 - 4 spread deviations out too much in some contrived cases
 - This was just my opinion
 - Some other power may be better, depending on what a "typical" model is for you.

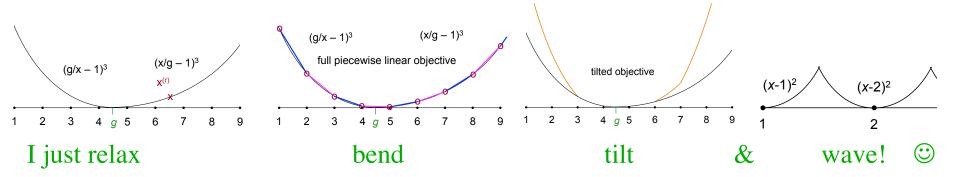


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Thanks

It takes a village to make me successful

- Thanks to Tim Tautges for suggesting and supporting this project
- Thanks to Tim Tautges, Rajeev Jain, for MeshKit
- Thanks to Carl Laird, for IPOPT
- Thanks to David Bommes for Graphics quads via mixed-integer quadratic programming
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- Is the government still shut down?
 Will mesh for food.

